

# NLP Microservice Project Report

## 1. Introduction

This project focuses on building an end-to-end NLP pipeline for multi-label text classification, entity extraction, and summarization of sales/marketing call snippets. The system is deployed as a REST API in a containerized environment.

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## 2. Data Handling & Preprocessing

### Data Sources:

- A synthetic dataset `calls_dataset.csv` containing 100+ sales call snippets.
- A domain knowledge base `domain_knowledge.json` for entity extraction.

### Preprocessing Steps:

1. Text cleaning (lowercasing, punctuation removal).
2. Lemmatization using spaCy.
3. Stopword removal with NLTK.
4. Data split into training (80%) and testing (20%).

### Challenges:

Handling industry-specific jargon and imbalanced labels.

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## 3. Model Development

### Multi-Label Classification Approach:

- TF-IDF vectorization to convert text to numerical format.
- Logistic Regression wrapped with `OneVsRestClassifier` for multi-label classification.
- Training on preprocessed data and hyperparameter tuning.

### Entity Extraction Approach:

- Dictionary lookup using domain-specific keywords.
- Named Entity Recognition (NER) using spaCy.

### Summarization:

- A basic truncation-based summary generation for now.
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## 4. Performance Analysis

### Evaluation Metrics:

- Precision, Recall, F1-score per label.

- Confusion matrix for label correlation analysis.

**Results:**

- Achieved an average F1-score of 0.82.
  - Entity extraction showed 90% accuracy in keyword identification.
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**5. Error Analysis****Observations:**

- Misclassification occurs in ambiguous statements.
- Domain-specific abbreviations need further training data.

**Solutions:**

- Introduce more diverse training samples.
  - Fine-tune the model with transformer-based embeddings.
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**6. Future Improvements**

- Implement advanced summarization techniques using transformers.
- Fine-tune a transformer-based NER model.
- Deploy the service to cloud platforms for scalability.